

Advanced Graphical Intelligence Logical Computing Environment (AGILE)

Modelling and Simulation

Dr. William Harrod | September 6, 2022



Intelligence Advanced Research Projects Activity

I A R P A

Creating Advantage through Research and Technology



AGILE Program



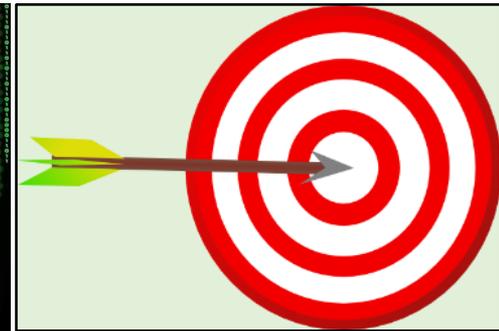
Open Designs



Co-design Process



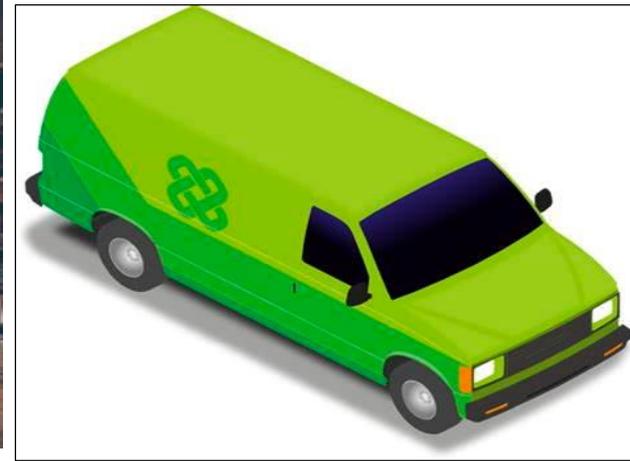
Modeling & Simulation



Target Metrics

- Create innovative computer architectures and designs that overcome the current and future data-analytics technical challenges.
- The program will result in the delivery of **system-level RTL designs** where the performance has been evaluated by using an application modeling and simulation environment.
- **Develop validated designs that achieve or exceed the AGILE Program Target Metrics. These results will be validated by an independent test and evaluation team.**

AGILE Program Details



Program Objectives:

- Enable data analytic problems that involve **10X** more data.
- Time to solution **10 - 100** times faster.

Research Effort:

- Develop validated designs that achieve or exceed the **AGILE Program Target Metrics**.
- These results will be validated by an independent test and evaluation team.

Deliverables:

- **Phase 1:** System-level functional model of architecture. Including runtime.
- **Phase 2:** Detailed (RTL) design for proposed **AGILE** system architecture, including runtime.

<https://www.iarpa.gov/research-programs/agile>

IARPA IS NO LONGER ACCEPTING PROPOSALS

Data Analytics Problem

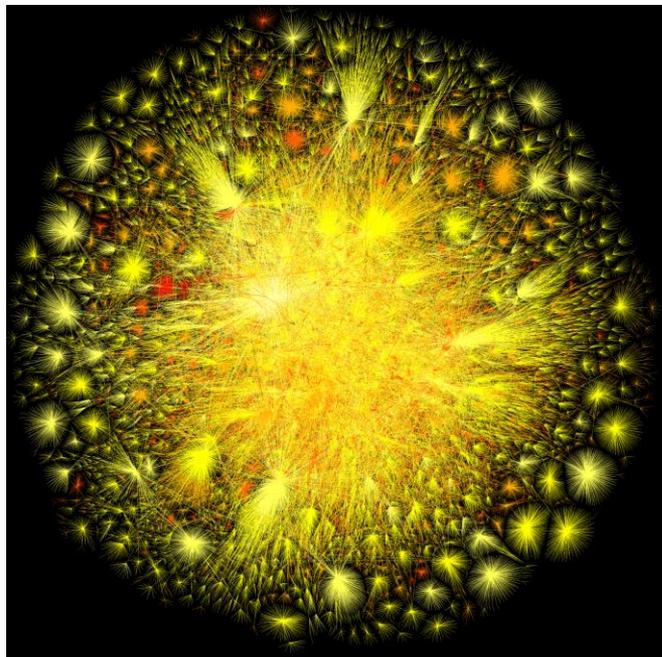


Entities are represented by vertices (V) with types and properties, and relationships are represented by edges (E) with types and properties.

The graphs are typically sparse
 --- that is $|E| \ll \ll |V|^2$

Graphs	Vertices	Edges
Social network	1 Billion	100 Billion
Internet	50 Billion	1 Trillion
Brain	100 Billion	100 Trillion

Technical Report NSA-RD-2013-056002v1,
 U.S. National Security Agency



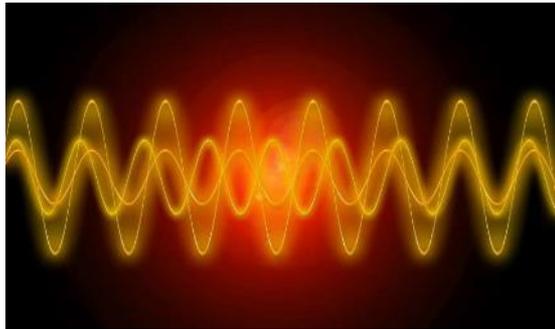
Internet Graph 2010

[The Opte Project](#)

Extracting Actionable Knowledge Methods

Graph Analytics	Machine Learning	Statistics Methods	Linear Algebra	Data Filtering
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“The variety and volume of data collected (today) ... far outpace the abilities of current systems to execute complex analytics ... and extract meaningful insights.” (Buono, D., *Computer*, August 2015)



Measurements
Sensors
Sound, Images, Video
Text



Floating Point
Integers
Strings
Bits



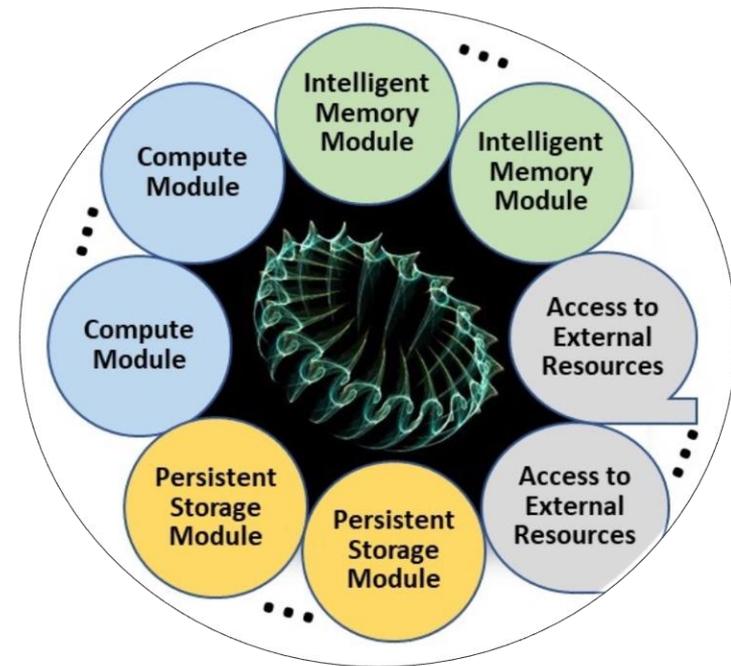
Graphs / Matrices
Documents
Tensors
Maps

- As problem sizes grow over time – data is increasingly sparse, random and diverse
- Increasing variety of data sources
- Vertex data & meta-data can vary in size from bytes to giga-bytes and larger



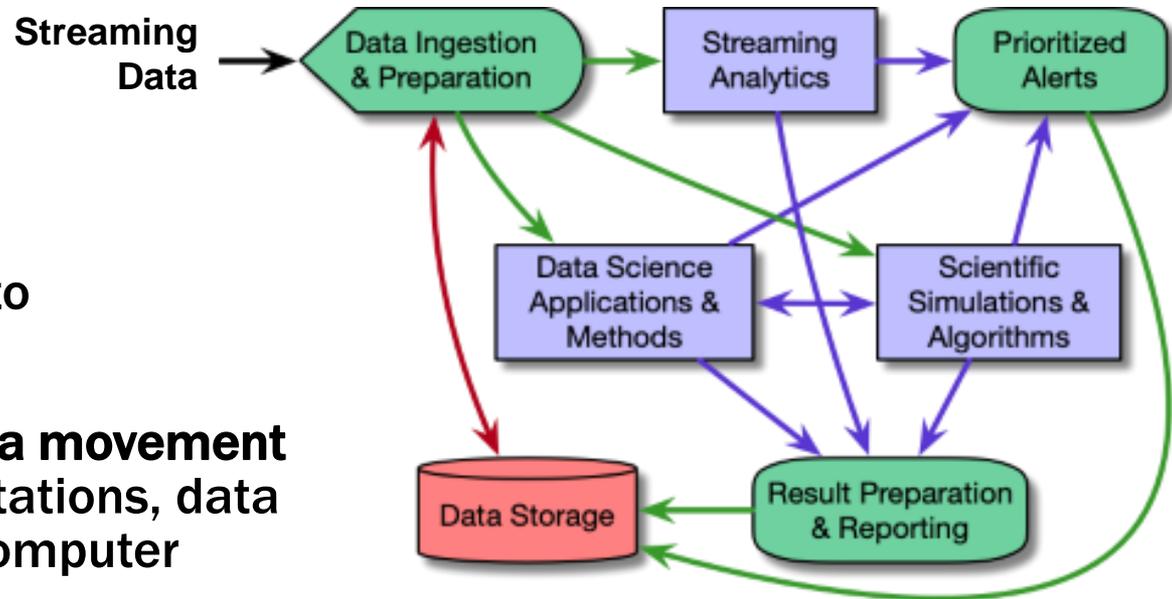
Innovative Research Strategies

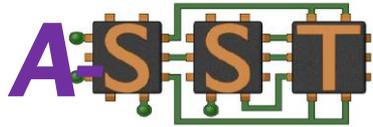
- **Tightly Integrated Subsystem Design**
communication, memory, compute & runtime
- **Data-driven compute elements**
including moving the compute to the data
- **Distributed memory management and security**
fine-grained addressing and protection of objects
- **System-level intelligent mechanisms**
large, random, time-varying data streams & structures
- **Global Name Space/Global Adaptive Data Transfer**
driven by complex workflow requirements
- **Dynamic adaptive runtime**
dynamically changing resource availability



AGILE Modular System Vision

- Results required in **near-real-time up to hours**
- Streaming data causes **unpredictable changes** to stored data
- **Extremely fine grain data movement and parallelism:** computations, data are distributed across computer
- Data computation tasks to be performed are typically **determined by the data and streaming queries**
- Tasks have **extremely poor data locality and data reuse**





Designs include a runtime system

- Performers are free to use any industry standard architectural development environment that they select.
- The Performers must supply their design specification using design-tool-neutral modeling or hardware description artifacts that can be imported to A-SST (SystemC, C/C++, SST, Verilog, or System Verilog).
- Performance evaluations and verification by the T&E Team will be conducted using A-SST and Firesim.



Attributes of AGILE Runtime



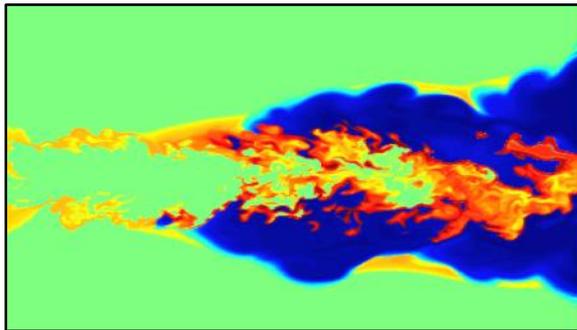
Runtime Model is an abstraction of computing system software structure and operation for a specific system model

Provides a conceptual framework for the co-design of technology: architecture, programming interfaces, and system software

Attributes:

- Extreme parallelism
- Asynchrony
- Self-discovered parallelism
- **Adaptive management**
- **Global name space**



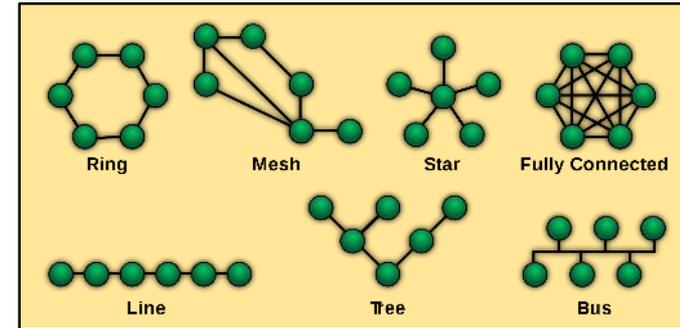


Multi-Physics Simulation

Jackie Chen, SNL



Accelerators



Conventional Networks

Designed for yesterday's applications

- Multi-physics simulations

Vendors are focused on incremental improvements

- Accelerators & supporting memory components
- Focused on processing not data challenges

Computers are not computational efficient or scalable for large scale graph analytics problems

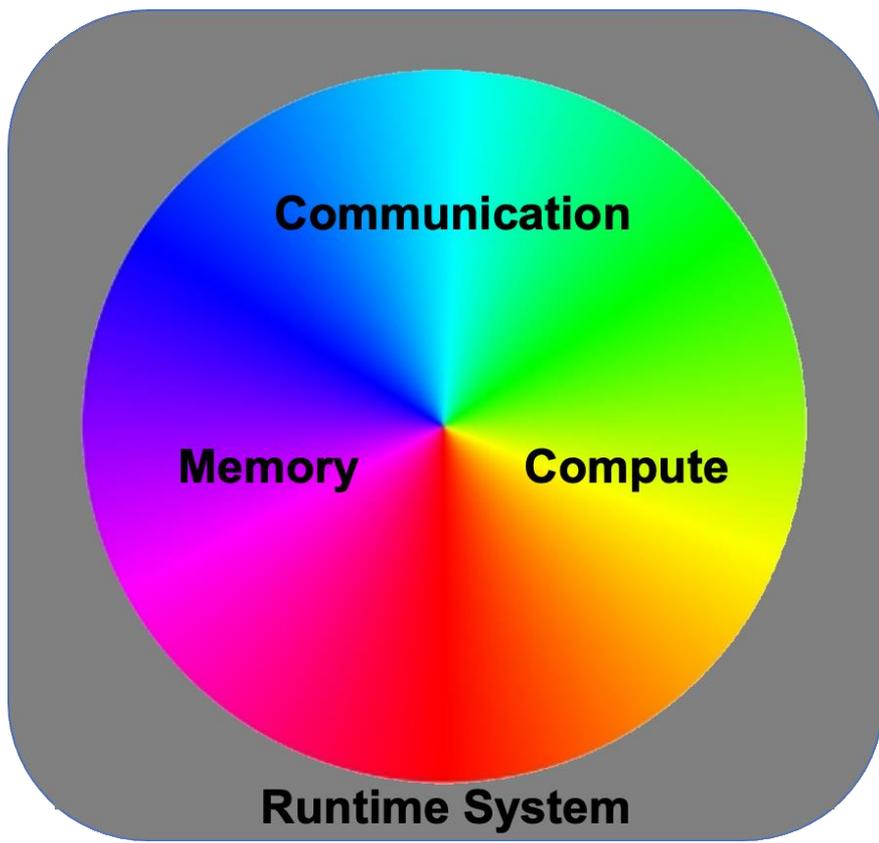
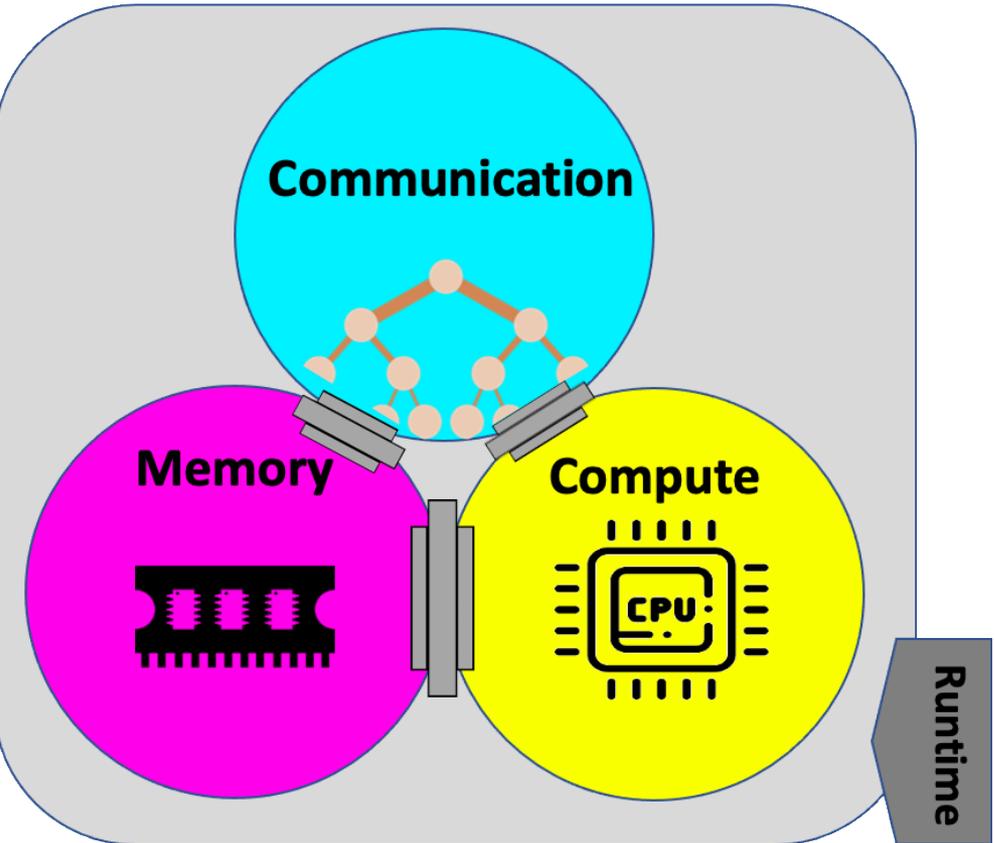
Over Provisioned Features

- Deep hierarchical memories
- Large-message interconnection networks
- Bulk, synchronous execution models



Today's Sub-optimal Computer Designs

AGILE Integrated Computer Designs



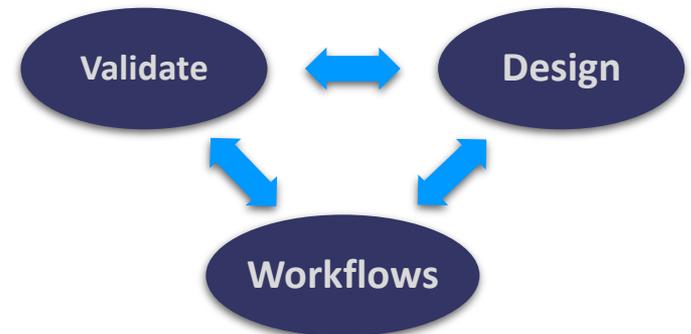
Workflows: Representative data analytic workflows, with realistic datasets

Design: AGILE Research Teams will perform research that will result in an integrated system design for data-driven computations

Utilize a co-design process involving : 1) AGILE workflows, benchmarks and kernels, 2) research/designs, and 3) modeling and simulation

Teams will develop detailed system-level designs, based on RTL designs and Functional models

Validate: AGILE enhanced modeling and simulation (A-SST) tool-kit will be utilized to evaluate a team's design characteristics and performance estimates



Output: Open system designs; all non-open IP must be licensable. System designs must enable system security and compliance.



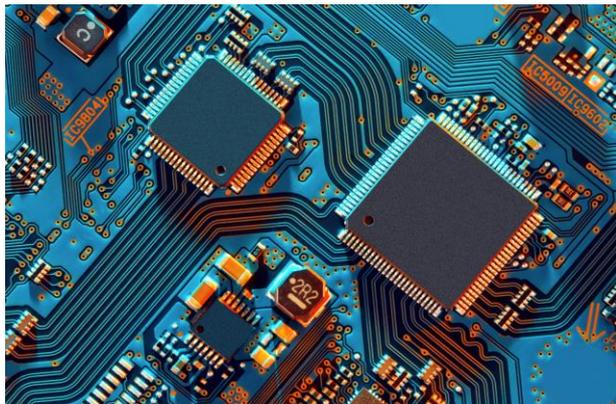
**Research
Teams**



**Test and
Evaluation
Team**

Independent Test and Evaluation Process Based on The Following Areas

Design V&V

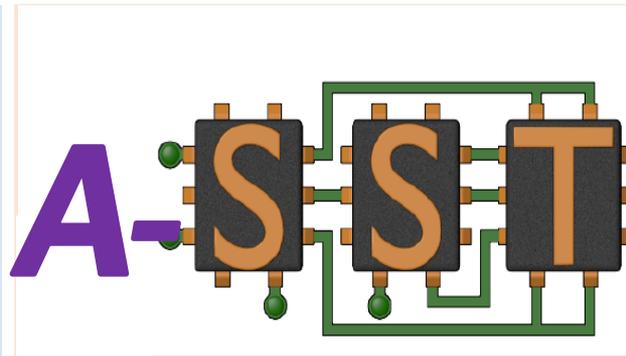


Validate Performers' Hardware & Application Test Plans

Evaluate Performers' models/designs for correctness & completeness

Validate the results generated using A-SST

ModSim



Validate Performers' models in the A-SST (Toolkit) & Firesim

Using A-SST, provides performance estimates of the Performers' models/designs

***AGILE-enhanced Structural Simulation Toolkit (Modeling and Simulation Environment)**

Application Codes



Develop AGILE Workflows and kernels

Baseline performance

Validate changes to the Performers' versions of the AGILE Workflows, kernels and benchmarks (optimized for their systems)



Multi-model Modeling & Simulation Methodology

Based on Sandia National Laboratories - Structural Simulation Toolkit (SST)

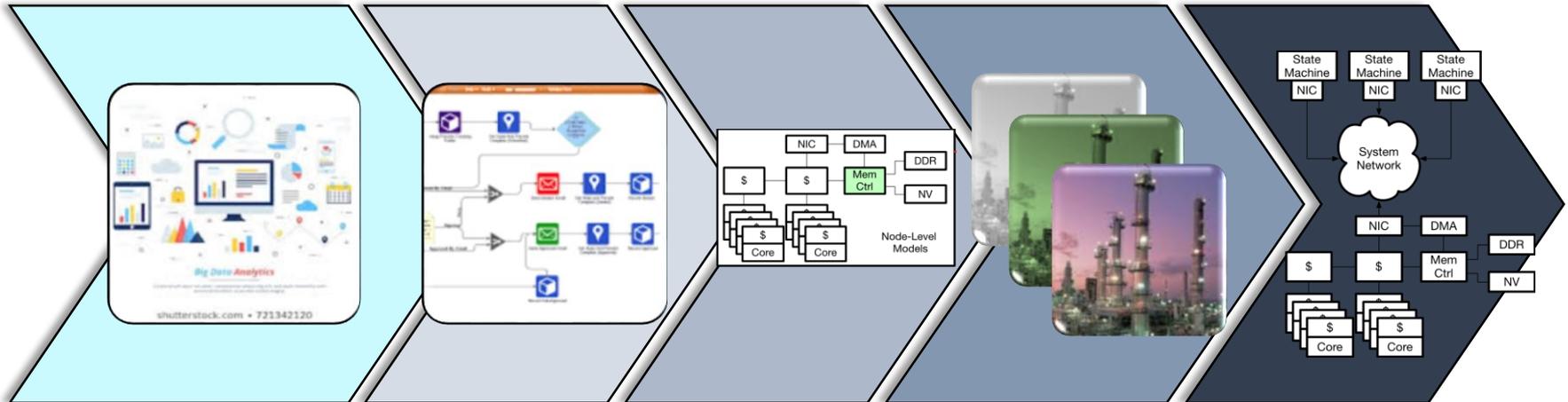
Application Representation

Workload-Defined Design Requirements

Execute High-Level Models

Refine / Optimize Designs

Execute Low-Level Designs



- Proxy of critical performance bottlenecks
- Capture workload

- Develop High-level Model
- Behavioral Model
- System – multi-node model

- Validate high-level bottlenecks
- Initial performance estimates and uncertainties

- Design low-level design experiments using initial high-level results
- Define ensembles for data-dependent workload sampling
- Develop multi-node design

- Refine performance models and model uncertainties
- Validate models against testbeds
- Improve design characteristics and performance



Workflows and Kernels



- Given the heterogeneity and complexity of data analytic workloads, kernels that measure individual metrics - FLOPS, TEPS, cache misses, network bandwidth - for a single data type cannot reflect the performance and scalability of full applications
- Only end-to-end workflows can reflect the performance and scalability of real-world analytic jobs
 - Ingestion, transformation, and storage of input data can take significant time, energy, and machine resources
 - Prioritization and display of output results can be costly
- Kernels are still valuable when measuring the speeds-and-feeds of individual system components and when systems/tools are too immature to run complete workflows





AGILE Applications

- Includes workflows, kernels and industry benchmarks
- Test programs / scripts
- Data sets or generators

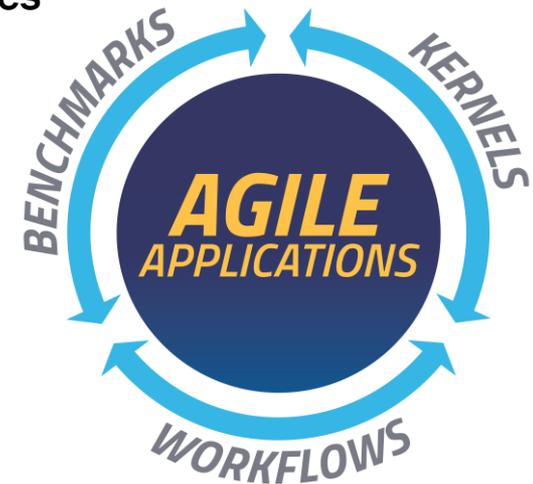
Reference codes will be written using SHAD

- Presents a shared-memory view of global memory
- STL-complaint, thread-safe, distributed data structures
- Concurrent insert/delete/modify and AMOs on all data structures
- Asynchronous data and task parallel programming constructs
- Multithreaded runtime that hides latencies (no data partitioning necessary)
- Runs on servers and clusters
- <https://github.com/pnnl/SHAD>

Algorithms can be substituted if they provide the same functionality



SCALABLE
HIGH-PERFORMANCE
ALGORITHMS &
DATA-STRUCTURES





Target Metric Tables

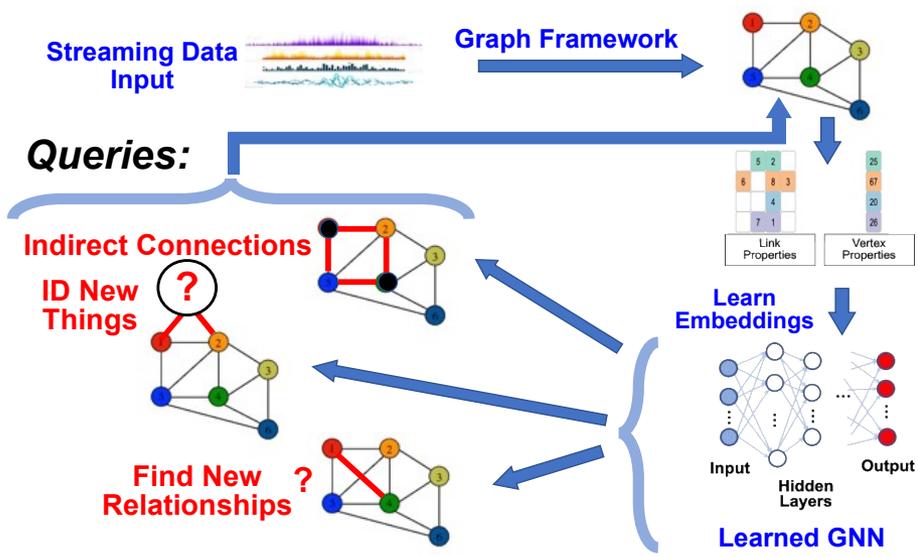
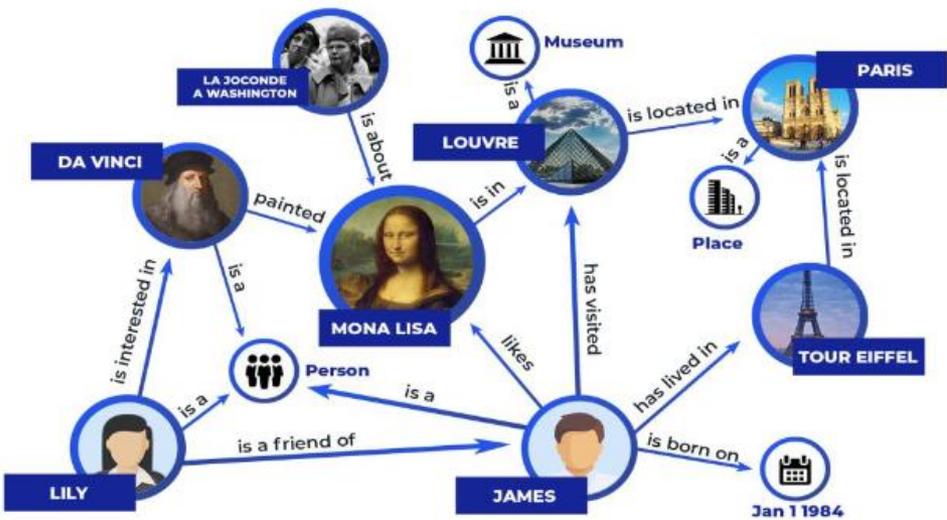


Knowledge		
Metric	Today	AGILE Target
Data ingestion rate	0.1 G data-elements per second	10 G data-elements per second from 3 or more sources
Time to learn embedding (Graph Size > 1 PB)	1,440 minutes	30 minutes
Time to classify vertices and edges	> 1,440 minutes	30 minutes
Time to predict and infer new relationship	> 1,440 minutes	30 minutes
Time to reason about higher-order relationships using multi-hop reasoning	1 – 2 hops (exact matches) in 30 minutes	3 – 5 hops (approximate/fuzzy matches) in 1 minute

Sequence Data		
Metric	Today	AGILE Target
Size of graph	0.01 PB ⁴	10 PB
Data ingestion rate	0.1 G data-elements per second from a single source, single data type	10 G data-elements per second from a three or more sources and data types
Insert/Delete/Modify rate for vertices and edges	0.01 G edits / second (batched)	10 G edits / second (continuous)
Pattern Detection per minute	Single event, linear paths, exact match	Multiple events, branches, prioritized approximate/fuzzy matching
Incremental analysis	NOT DONE	Commensurate with data rate
Time to complete multiple day / multiple location queries	NOT DONE	Completed in minutes

Detection		
Metric	Today	AGILE Target
Size of graph	0.01 PB ⁴	10 PB
Data ingestion rate	0.1 G data-elements per second from a single source, single data type	10 G data-elements per second from a three or more sources and data types
Insert/Delete/Modify rate for vertices and edges	0.01 G edits / second (batched)	10 G edits / second (continuous)
Pattern Detection per minute	Single event, linear paths, exact match	Multiple events, branches, prioritized approximate/fuzzy matching
Incremental analysis	NOT DONE	Commensurate with data rate
Time to complete multiple day / multiple location queries	NOT DONE	Completed in minutes

Network		
Metric	Today	AGILE Target
Construct 1 PB graph through game theoretic modeling	120 minutes	2 minutes (60x faster)
Identification of top k influential nodes (simple model)	60 minutes	1 minute (60x faster)
Identification of top k influential nodes (enhanced model)	600 minutes	30 minutes (20x faster)
Propagate labels/confidence score	120 minutes	2 minutes (60x faster)
Incremental analysis	NOT DONE	Never recomputed from scratch



What is it:

- A semantic network of persons, places, objects, events, situations, or concepts, and the relationships among them
- Integrates multiple data sources with disparate types of entities (vertices) and relationships (edges)
- Ontologies are used to establish a logical, hierarchy of types creating a formal representation of the entities in the graph

Knowledge graph use cases:

- Discover new entities, relationships & facts
- Explain the contextual reasons for a particular event
- Explain why a human expert should look at emerging event
- Answer complex questions that are beyond database queries

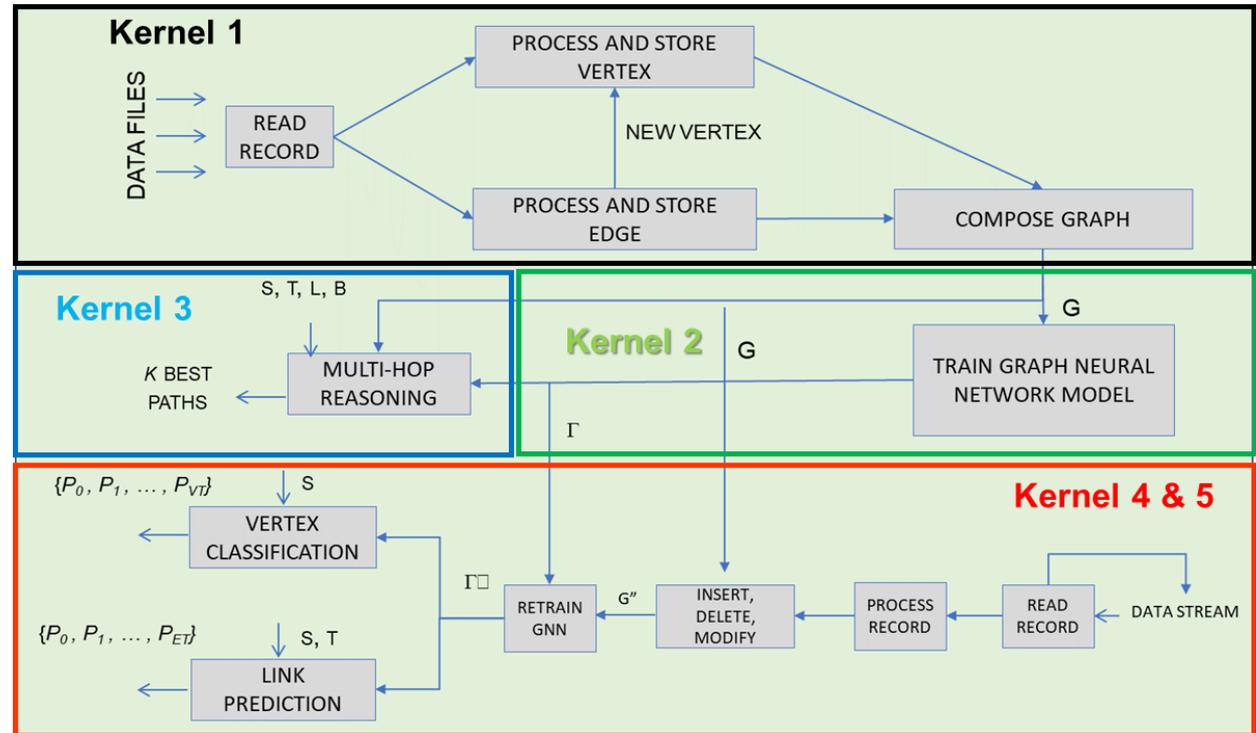
Workflow 1 – Kernels

- Kernel 1: measures streaming data ingestion rate, the time to read data records, transform the raw data, resolve vertex and edge ambiguities and build-out the common internal data structures used by downstream tasks.

- Multi-hop Reasoning *Indirect connections* given vertices s and t in G , return the “best” k paths from s to t
Kernel 3

- Vertex Classification *ID new things* given unlabeled v in G with properties (p_1, \dots, p_n) and incident edges $\{e_1, \dots, e_k\}$, return the type of v
Kernel 4

- Link Prediction *Find new relationships* given s, t in G such that edge $\{s, t\}$ does not exist in G , predict the existence and edge type of $\{s, t\}$
Kernel 5



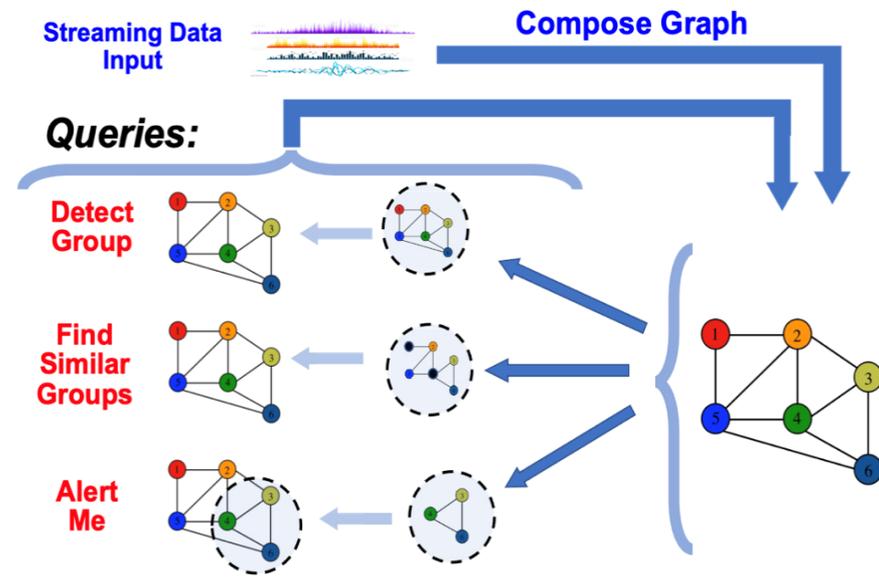
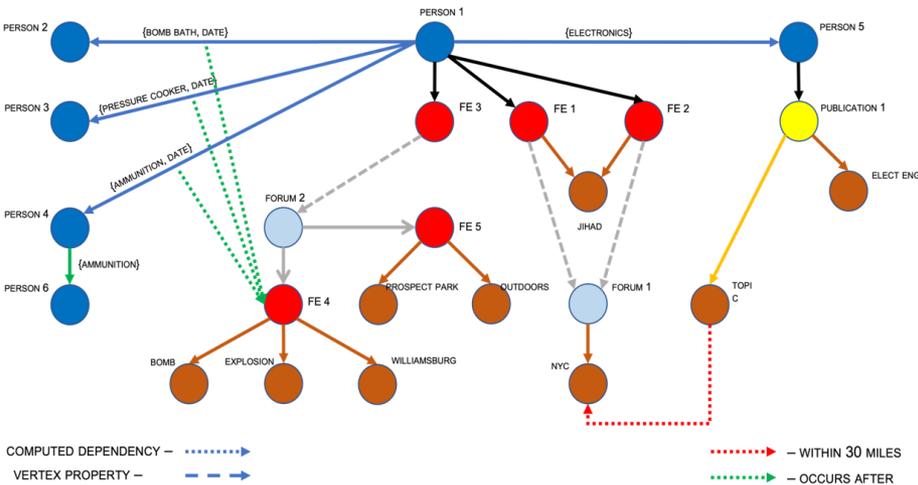
Knowledge Graph Target Metrics

Kernel 1 – Row 1 Kernel 4 – Row 3

Kernel 2 – Row 2 Kernel 5 – Row 4

Kernel 3 – Row 5

Kernels 4, 5 involve updating model from Kernel 2



What it is:

Perform exact, approximate, and partial matching of a pattern graph against a world graph.

Let $G = (V, E, C_V, C_E)$ be a property graph where V is a set of vertices, E is a set of edges, C_V is the set of vertex property labels, and C_E is the set of edge property labels. Let P be a pattern graph and let $\{T_1, T_2, \dots, T_K\}$ be K subgraphs of P such that their union is P .

Pattern detection use cases:

- find similar images
- find organizations
- monitor for specific pattern

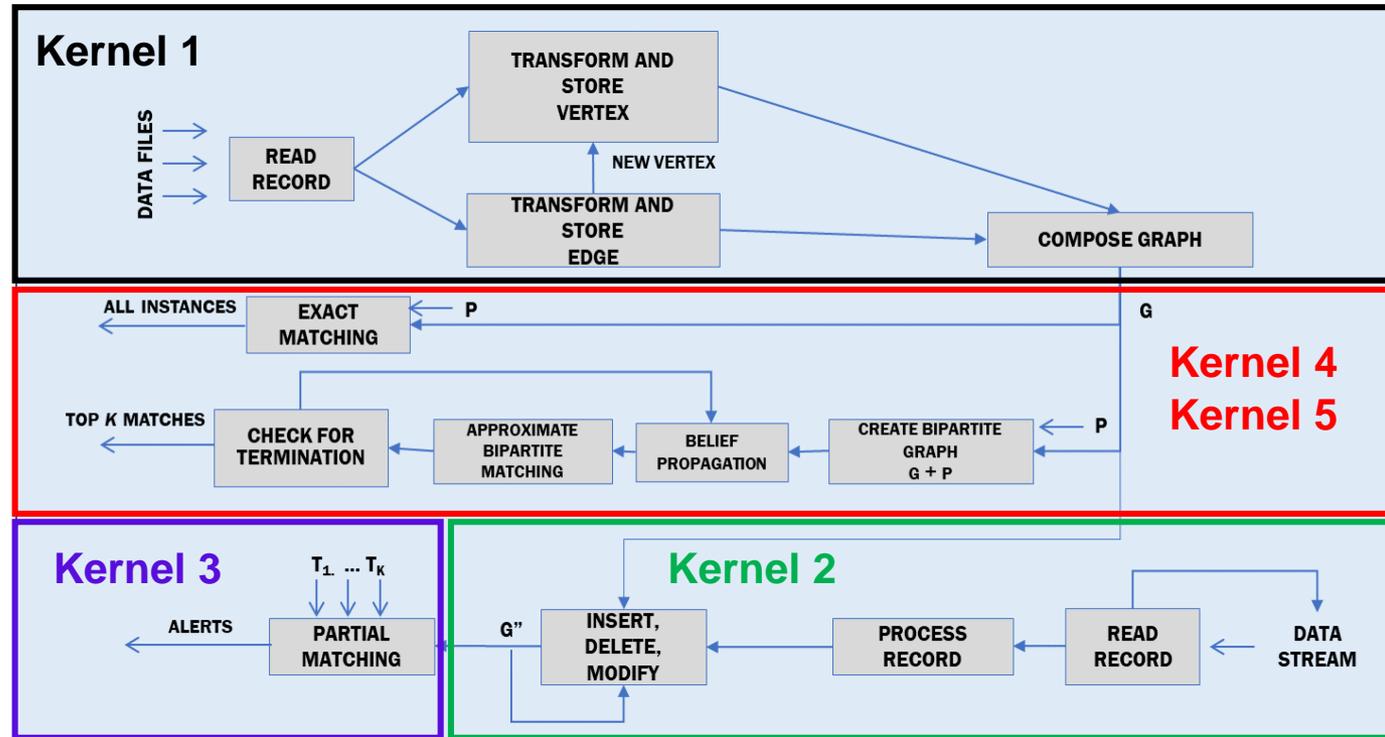
Workflow 2 – Kernels

- Kernel 1:** measures streaming data ingestion rate, the time to read data records, transform the raw data, resolve vertex and edge ambiguities and build-out the common internal data structures used by downstream tasks.

- Exact Match**
Detect Groups
 Find all instances of **P** in **G**
Kernel 4

- Approximate Matching**
Find Similar Groups
 Return the *N* closest matches of **P** in **G** as measured by some graph edit function.
Kernel 5

- Partial Matching**
Alert Me
 As new data is added to **G**, alert when a subgraph T_i appears in **G**.
Kernel 2 & 3



Pattern Detection Target Metrics
 Kernel 1 – Row 1 **Kernel 4 – Row 3**
 Kernel 2 – Row 4 **Kernel 5 – Row 3**
 Kernel 3 – Row 4



Workflows and Benchmarks

Objectives

- Enable data analytic problems that involve **10X** more data
 - Time to solution **10 times faster**
 - Designs must be able to achieve the **Target Metrics**
- These objectives motivate the **Workflow-driven R&D plans** specified in the proposals
 - Performance estimates will be validated by **T&E Team using A-SST (based on Sandia's SST)**

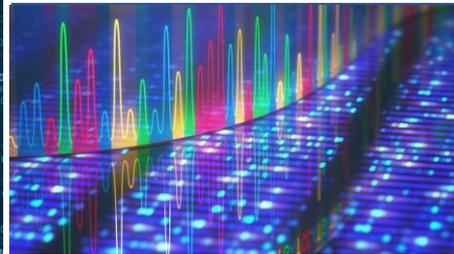
Workflows



Knowledge Graphs



Pattern Detection



Sequence Data



Network

- AGILE will provide reference implementations - Performers can modify to optimize for their design - Performers will use in the Co-Design process**
- Develop and investigate design that can achieve the target metrics for the Workflows**

Benchmarks

Breadth First Search

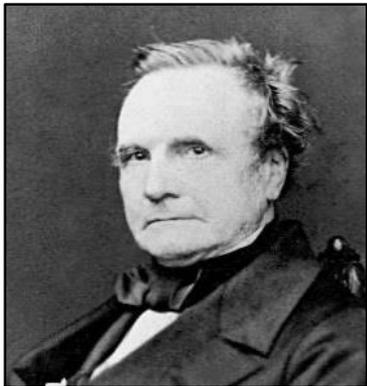
Counting Triangles

Jaccard Similarity

- Performers will utilize the **Benchmark Codes** in the **Co-Design process**
- Develop and investigate design that can **achieve the target metrics** for the **Benchmarks**



Looking for the next computer pioneer, whose picture will appear below



Charles Babbage
Mechanical Computer



Alan Turing
Turing Machine



John von Neumann
von Neumann Arch.



John Atanasoff
1st Electric Computer



Grace Hopper
Programming

- Architectures that enable scalable, efficient execution of data intensive applications
- System-level intelligent mechanisms for moving, accessing and storing large, random, time-varying data streams, structures, objects, and knowledge
- Dynamic adaptive runtime systems to match activity demands to changing resource availability supported by hardware capabilities
- Declarative interfaces for intelligent programming environments that provide an intelligent determination of efficient and scalable data and computation operations



ModSim/Firesim Challenges



Demonstrating that the designs can achieve or exceed target metrics

Modeling a system level design when executing appropriately sized applications with realistic data sets

Runtime system –

- Required by the design – evaluated using A-SST
- Developed on conventional platform – evaluated on baseline platform

Complete the evaluation in a reasonable amount of time

Verifying the design

Evaluating security





Summary



AGILE Program has three tier evaluation process:

- 1. End-to-end applications (Workflows)**
that measure full system performance
 - Data sets at different scales
 - Data ingestion and preparation
 - Multiple computational components
- 2. Kernels derived from data-intensive applications**
- 3. Industry standard benchmarks**
 - Breadth-first search
 - Triangle counting
 - Jaccard coefficient

AGILE utilizes ModSim to estimate and evaluate the performance of the designs, when executing the AGILE Applications